

(K,L) eigenvectors in max-min algebra

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Fuzzy Sets and Systems

(K,L)-eigenvectors in max-min algebra

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Using the concept of (K,L) -eigenvector, we investigate the structure of the max-min eigenspace associated with a given eigenvalue of a matrix in the max-min algebra (also known as fuzzy algebra). In our approach, the max-min eigenspace is split into several regions according to the order relations between the eigenvalue and the components of x . The resulting theory of (K,L) -eigenvectors, being based on the fundamental results of Gondran and Minoux, allows to describe the whole max-min eigenspace explicitly and in more detail.

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(K, L)-eigenvectors in max-min algebra

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Abstract

21 Using the concept of (K, L) -eigenvector, we investigate the structure of the max-
22 min eigenspace associated with a given eigenvalue of a matrix in the max-min
23 algebra (also known as fuzzy algebra). In our approach, the max-min eigenspace
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25 value and the components of x . The resulting theory of (K, L) -eigenvectors,
26 being based on the fundamental results of Gondran and Minoux, allows to de-
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28 *Keywords:* max-min, fuzzy algebra, eigenvector

29 *AMS Classification:* 15A80, 15A18

1. Introduction

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36 By max-min algebra we mean the unit interval $\mathbb{B} = \langle 0, 1 \rangle$ equipped with
37 the arithmetic operations of "addition" $a \oplus b = \max(a, b)$ and "multiplication"
38 $a \otimes b = \min(a, b)$. Algebraically speaking, max-min algebra is a semiring where
39 both arithmetic operations are idempotent. Let us also note that, algebraically,
40 max-min algebra is an example of incline algebra of [4].

41 The arithmetic operations \oplus and \otimes can be extended to matrices and vectors
42 in the usual way so that for any matrices $A = (a_{ij})$ and $B = (b_{ij})$ of appropriate
43 dimensions we can define their "sum" $A \oplus B$ and "product" $A \otimes B$ by the usual
44 rules: $(A \oplus B)_{ij} = a_{ij} \oplus b_{ij}$ and $(A \otimes B)_{ij} = \bigoplus_k a_{ik} \otimes b_{kj}$. For a square matrix
45 $A \in \mathbb{B}^{n \times n}$ we can also define its max-min matrix powers: $A^k = \underbrace{A \otimes \dots \otimes A}_k$,
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48 where k is a natural number. Note that $A^0 = I$, the usual identity matrix.
49 Further we will systematically omit the product sign \otimes , for brevity.

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As usual in tropical/idempotent algebras, to each matrix $A \in \mathbb{B}^{n \times n}$ we associate a digraph $\mathcal{D}(A) = (N, E)$ with the node set $N = \{1, \dots, n\}$ and edge set $E = \{(i, j) : a_{ij} > 0\}$. Each edge has a weight a_{ij} . A sequence of edges $(i_0, i_1), (i_1, i_2), \dots, (i_{k-1}, i_k)$ where each edge belongs to E is called a walk whose length is k and whose weight is given by the max-min product $a_{i_0 i_1} \dots a_{i_{k-1} i_k}$.

One of the motivations to study max-min algebra comes from the theory of fuzzy sets where the operation $\min(a, b)$ is one of the most useful examples of triangular norms, see Klement, Mesiar and Pap [19]. See also Gondran and Minoux [17] for more on semirings and other algebraic models relevant to the theory of fuzzy sets.

The main goal of the present paper is to further investigate the structure of the max-min eigenspace associated with a given eigenvalue. For a given matrix $A \in \mathbb{B}^{n \times n}$ and a number $\lambda \in \mathbb{B}$, the max-min eigenspace associated with λ is the set of vectors $x \in \mathbb{B}^{n \times 1}$ (called λ -eigenvectors) such that

$$Ax = \lambda x. \tag{1}$$

Note that it is indeed a space in the sense of max-min algebra: for any $\alpha, \beta \in \mathbb{B}$ and any u and v satisfying (1) $\alpha u \oplus \beta v$ also satisfies this equation. Let us now give some examples to motivate the study of this eigenproblem and our approach to it.

Example 1.1 (Medical symptoms and therapy). One of the first models, in which the max-min eigenvectors were used, was concerned with searching invariants in therapeutic recommendations. It was put forward by Sanchez [27], see also a monograph by Rakus-Andersson [26] for a more recent account of this method and more references. In this model, the matrix $A \in \mathbb{B}^{n \times n}$ expresses a fuzzy relation describing the relative success of treating n symptoms of some illness by some drug. In this matrix, a diagonal entry a_{ii} is equal to the recovery rate from the i th symptom, and entry a_{ij} , for any $i \neq j$ is, the rate at which either both symptoms i and j are absent after the application of drug, or symptom i is absent and symptom j is still present. In other words, a_{ij} is the rate at which “the action of the drug is equal or stronger on the i th symptom than on the j th symptom” [26], designed to account for the link between the symptoms and action of the drug on them. The max-min eigenvectors x associated with eigenvalue 1 appear as *eigen fuzzy sets*. Note that in this application, one is usually interested in the greatest left eigenvector of A , i.e., the greatest x such that $xA = x$. This vector, together with some other vector, is then used to estimate reliable intervals for the rate at which a drug removes the n symptoms.

The greatest max-min eigenvector is easy to use, as it can be quickly found by a simple iterative procedure [28, 26]. However, it can be seen as giving an overly optimistic judgement, and in this situation one may be interested 1) to impose a condition that the recovery rate is no bigger than λ and 2) to require that certain components of vector x , while being below λ , stay at their initial level x_i . A further development of this idea in the context of medical treatments is beyond the scope of the present paper.

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9 The greatest left max-min eigenvectors of a given matrix A with entries in
10 the interval $[0, 1]$ were also used by Nobuhara et al. [25] in image reconstruction,
11 where a given image was encoded by means of max-min and min-max eigenvectors
12 of A , as well as fixed points of various convex combinations of the max-min
13 and min-max products associated with A , and then successfully reconstructed
14 using them.

15
16 **Example 1.2 (Security in a computer framework).** This example was put
17 forward as a motivation for studying strongly tolerant interval eigenvectors in
18 max-min algebra [13]. A computer network consisting of servers S_1, \dots, S_n ,
19 data storage units D_1, \dots, D_n and a logical unit L is considered. Lines K_{ij} ,
20 for $i, j \in I$, connect S_i with D_j , while lines L_j connect every D_j with L . The
21 security level of each line in the network is measured by values in the real interval
22 $[0, 1]$, where value 0 stands for completely unsecure connection and value 1
23 stands for completely secure connection. The security of every K_{ij} (data security)
24 is denoted by a_{ij} , and the security of L_j (logic security) is denoted by
25 x_j . The maximal security level y_i of information transfer from servers S_i to the
26 logical unit L through the given storage units is the i th component of the vector
27 $y = A \otimes x$. If we wish to keep the data security unchanged by the information
28 transfer from servers S_i to the logical unit L through the given storage units,
29 then $y = x$ should hold. Moreover, we have to consider the security level of the
30 system itself, which depends on the accessible technologies and the available
31 budget. The technological security of the system can decrease the security of
32 the processed data below the level λ , for some $\lambda \in \langle 0, 1 \rangle$, and then λx becomes
33 new target level of security. Note that $A(\lambda x) = \lambda x$ in max-min algebra, which
34 means that the invariance of the new levels of security will be ensured. If we
35 additionally require that the new levels of security should become exactly equal
36 to λ in some components and stay equal to x_i in the others, then we are led to
37 study (K, L) -eigenvectors, introduced and studied in the present paper.
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41 Problem (1) has been studied in max-min algebra at least since Gondran and
42 Minoux [15, 16] and Sanchez [27]. The approach taken in [15, 16] resembles that
43 of max-plus algebra, where eigenspaces are characterized as particular subspaces
44 of the column span of Kleene star. There is also a number of works where a
45 different approach is taken. Sanchez [27, 28] is focused on the largest eigenvector
46 associated with the eigenvalue 1 and suggests an algorithm that can be used to
47 compute it in practice. In the same vein, Cechlárová [6] describes lower and
48 upper basic eigenvectors in terms of the associated graph. The structure of the
49 eigenspace of increasing max-min eigenvectors $x_1 \leq x_2 \leq \dots \leq x_n$ in max-
50 min algebra is described by Gavalec [10], and various types of max-min interval
51 eigenvectors have been studied in Gavalec et al. [14].

52
53 Gondran and Minoux obtained fundamental results for (1) also over more
54 general semirings with idempotent multiplication, see [18][Section 6.3] for one
55 of the latest accounts. We are going to use these results. However, we observe
56 that the theory as presented in that monograph describes only the solutions
57 whose all components are less than or equal to λ [18][Ch. 6, Corollary 3.5].
58

Several examples when (1) also admits other solutions can be found in the present paper. To describe those other solutions we adopt an approach which is similar to that of Gavalec et al. [12], where the eigenproblem in max-Lukasiewicz algebra was studied (see also Gavalec and Němcová [11]). Namely, given A and λ we consider a partition of $N = \{1, \dots, n\}$ into two disjoint subsets K and L such that $K \cup L = N$ and pose a problem of describing all λ -eigenvectors x that satisfy $x_i \leq \lambda$ and hence $\lambda x_i = x_i$ for all $i \in K$ and $x_i \geq \lambda$ and hence $\lambda x_i = \lambda$ for all $i \in L$. When $K = N$ we call such vectors "principal λ -eigenvectors" since (1) becomes $Ax = x$, and when $L = N$ we call such vectors "background λ -eigenvectors", in analogy with [12]. In the latter case (1) becomes $Ax = \lambda \mathbf{1}$, where $\mathbf{1}$ denotes the vector of all 1's. principal eigenvectors were described in [18][Ch. 6, Corollary 3.5], which we revisit here in Corollary 3.2. Background eigenvectors are easy to obtain: see Proposition 3.1 and Proposition 3.2 below. Principal and background λ -eigenvectors are fundamental for describing the (K, L) -eigenvectors associated with λ in the case of general K, L and λ . Their description is stated in Theorem 3.1, which can be considered as our main result.

Theorem 3.1 also yields a method for constructing a generating set for the whole λ -eigenspace, although the computational effort may grow exponentially with matrix size. First of all, note that, for general K and L , the set of (K, L) -eigenvectors is not a max-min space any more, since the set $\{x: x_i \geq \lambda\}$ is not a max-min space. However, it is a *max-min convex set*: if x and y are (K, L) -eigenvectors associated with λ and $\alpha \oplus \beta = 1$, then $\alpha x \oplus \beta y$ is also such a (K, L) -eigenvector. This follows since (like in the usual convexity) any max-min space is a max-min convex set, and the sets defined by max-min affine inequalities such as $x_i \leq \lambda$ or $x_i \geq \lambda$ are also max-min convex. For arbitrary vectors $x^1, \dots, x^n \in \mathbb{B}^n$ one can consider their *max-min convex combinations*: $\alpha_1 x^1 \oplus \dots \oplus \alpha_n x^n$ where $\alpha_1 \oplus \dots \oplus \alpha_n = 1$ is required, and the set of all such combinations (for given x^1, \dots, x^n) is called the *max-min convex hull* of these points. Some max-min convex sets are max-min convex hulls of a finite number of points. The easiest and most important example is \mathbb{B}^n , which is the max-min convex hull of the zero vector $\mathbf{0}$ and unit vectors e^i , whose i th component is equal to 1 and the rest of the components are equal to 0. Theorem 3.1 will provide for a comprehensive description of any set of (K, L) -eigenvectors as a max-min convex hull of a finite number of points. Then we can take the union of these sets and obtain a set of points whose max-min convex hull is the whole λ -eigenspace. Taking out the zero vector $\mathbf{0}$, we then obtain a generating set for the whole λ -eigenspace. See, e.g., Nitica and Sergeev [23, 24] for more on max-min convexity.

All new results of this paper are obtained in Section 3. Preliminary notions and results from Gondran and Minoux [18] and Butkovič et al. [3], which provide the necessary algebraic tools, are given in Section 2.

2. Some problems of max-min algebra

In this section we will give some necessary notions and facts from max-min algebra on which our study of max-min (K, L) -eigenvectors will be based.

We will start with defining the notions of metric matrix and Kleene star and (following Gondran and Minoux [18]) giving a description of the set of principal eigenvectors (x such that $Ax = x$). This will be followed by describing the solution set to max-min Bellman (Z -matrix) equation (following Butkovič et al. [3] or Krivulin [21]) and solving a special type of max-min equation (8) in Subsection 2.2.

2.1. Metric matrix, Kleene star and the principal eigenproblem

For a square matrix $A \in \mathbb{B}^{n \times n}$ let us define its metric matrix $A^+ = (a_{ij}^+)_{i,j=1}^n$ and Kleene star $(a_{ij}^*)_{i,j=1}^n$ by the following series:

$$\begin{aligned} A^+ &= A \oplus A^2 \oplus A^3 \oplus \dots \\ A^* &= I \oplus A \oplus A^2 \oplus \dots \end{aligned} \tag{2}$$

where I denotes the identity matrix, whose diagonal entries are equal to 1 and off-diagonal entries are 0. It is well-known that in max-min algebra these series always converge and, moreover, can be truncated:

$$\begin{aligned} A^+ &= A \oplus \dots \oplus A^n, \\ A^* &= I \oplus A \oplus \dots \oplus A^{n-1}. \end{aligned} \tag{3}$$

In terms of the associated graph, a_{ij}^+ , being equal to a_{ij}^* when $i \neq j$, is the maximal (max-min) weight of all walks connecting i to j with unrestricted length. So is the *optimal walk interpretation* of metric matrices and Kleene stars.

Metric matrices, Kleene stars and associated digraphs provide some of the basic tools for the max-min eigenproblem. Let us start with the *principal eigenproblem*: the problem of identifying all vectors x that satisfy $Ax = x$ for a given matrix A . Such vectors are called *principal eigenvectors* of A .

For each principal eigenvector x , following the terminology of [1], define its *saturation graph* $\text{Sat}(A, x)$ as the graph consisting of all edges (i, j) that satisfy $a_{ij} \otimes x_j = x_i$ and all nodes on these edges. This graph in general has several maximal strongly connected components, and let $C(A, x)$ denote a subset of $N = \{1, \dots, n\}$ that contains one node from each strongly connected component of $\text{Sat}(A, x)$. We now state a description of the set of principal eigenvectors, which is due to [18].

Theorem 2.1 (Gondran-Minoux [18] Section 6.3). *The set of principal eigenvectors is a max-min space generated by vectors*

$$a_{ii}^+(A^*)_{.i}, \quad i = 1, \dots, n. \tag{4}$$

More precisely, each vector of (4) is a principal eigenvector, and each principal eigenvector x can be represented as

$$x = \bigoplus_{i \in C(A, x)} x_i a_{ii}^+(A^*)_{.i}. \tag{5}$$

We will also use the theory of algebraic Bellman equation

$$x = Ax \oplus b, \quad (6)$$

studied over general semirings, e.g., in Carré [5], Litvinov and Maslov [22]. In nonnegative linear algebra this equation is also known as Z -matrix equation [3].

Although (6) has been known for decades, the following fundamental result, describing the whole solution set to (6), was formulated only recently in Butkovič et al. and Krivulin [3, 20, 21].

Theorem 2.2. *Let $A \in \mathbb{B}^{n \times n}$ and $b \in \mathbb{B}^n$. Equation (6) always has solutions, and the set of these solutions is*

$$\{A^*b \oplus v : Av = v\}. \quad (7)$$

2.2. Special type of equation

We also need to describe the solution set for the system

$$Az \oplus b = \lambda \mathbf{1}. \quad (8)$$

where $A \in \mathbb{B}^{m \times n}$, $z \in \mathbb{B}^n$ and $b \in \mathbb{B}^m$ (for arbitrary natural numbers m and n).

We will study this system under the condition that all coefficients of A and b are less than or equal to λ :

$$a_{ij} \leq \lambda, \quad b_i \leq \lambda \quad \forall i, j \in N. \quad (9)$$

The description will be obtained in terms of coverings and minimal coverings, following the known solution method for systems of the form $A \otimes x = b$ and $A \otimes x \geq b$ in max-plus and max-min algebra (see, e.g., Butkovič [2], Cunningham-Green and Cechlárová [8], and Elbassioni [9]).

Note that if the system (8) is solvable, then in every row of the equation, we have to obtain λ either from Az or from b .

Let us denote $I_0 = \{i : b_i < \lambda\}$ and $C_j = \{i \in I_0 : a_{ij} = \lambda\}$ for $j \in N$.

Furthermore, for $W \subseteq N$ we denote C^W and z^W by putting

$$C^W = \{C_j : j \in W\} \quad (10)$$

$$z_j^W = \begin{cases} \lambda & \text{if } j \in W \\ 0 & \text{otherwise} \end{cases} \quad (11)$$

We say that C^W is a *covering* of I_0 , if $I_0 = \bigcup C^W (= \bigcup_{j \in W} C_j)$. Moreover, C^W is a *minimal covering* of I_0 , if

$$I_0 \neq \bigcup C^V \quad \text{for every } V \subseteq W, V \neq W. \quad (12)$$

Proposition 2.1. *$z \in \mathbb{B}^n$ is a solution of (8) with condition (9) if and only if $z \geq z^W$ for some $W \subseteq N$ such that C^W is a minimal covering of I_0 .*

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PROOF: Suppose $z \in \mathbb{B}^n$ is a solution of (8). Then $b_i < \lambda$ implies $(A_i)z = \lambda$. Then, also using (9), we obtain that for every $i \in I_0$ there is $j = j(i) \in N$ with $a_{ij(i)} = \lambda$ and $z_{j(i)} \geq \lambda$. Denoting $W = \{j(i) : i \in I_0\}$, we get $I_0 = \bigcup C^W$ and $z \geq z^W$. Without loss of generality, we can assume that the covering C^W of I_0 is a minimal one.

For the converse implication, suppose $z \geq z^W$ for some $W \subseteq N$, with C^W being a minimal covering of I_0 . Then, for any $i \in I_0$, there is $j = j(i) \in W$ such that $i \in C_{j(i)}$. That is, $z_{j(i)} \geq z_{j(i)}^W = \lambda$ and $a_{ij(i)} = \lambda$, which gives $(A_i)z = \lambda$. \square

Corollary 2.1. *Equation (8) with condition (9) is solvable if and only if there exists a covering C^W of I_0 . In this case, solution set to (8) can be represented as $\bigcup_W S^W$ where the union is taken over $W \subseteq N$ such that C^W is a minimal covering of I_0 and*

$$\begin{aligned} S^W &:= \{z : z^W \leq z \leq \mathbf{1}\} \\ &= \{z : \lambda \leq z_k \leq 1 \text{ for } k \in W, z_k \geq 0 \text{ for } k \in N \setminus W\}, \end{aligned} \quad (13)$$

Note that if $I_0 = \emptyset$ then the minimal covering of I_0 is C^W with $W = \emptyset$ and the unique minimal (and hence the least) solution of (8) is $\mathbf{0}$ and the solution set is $\{z : \mathbf{0} \leq z \leq \mathbf{1}\}$.

Set $S^W = \{z : z^W \leq z \leq \mathbf{1}\}$ can be also algebraically expressed as follows.

$$S^W = \{z^W \oplus v_N : v_N \in \mathbb{B}^{|N|}\} \quad (14)$$

Example 2.1. We shall illustrate the solution of (8) by the following system:

$$A = \begin{pmatrix} .3 & .5 & .3 \\ .6 & .6 & .2 \\ .6 & .3 & .6 \end{pmatrix}, \quad b = \begin{pmatrix} .6 \\ .3 \\ .2 \end{pmatrix}$$

and consider $\lambda = .6$.

Then the set $I_0 = \{2, 3\}$. From the entries of A we can derive sets $C_1 = \{2, 3\}$, $C_2 = \{2\}$ and $C_3 = \{3\}$.

For $W = \{1\}$, we have $C^W = \{C_1\}$. This set is a minimal covering of I_0 , the minimal solution here is $z^W = z^{\{1\}} = (.6, 0, 0)$ and every $z \geq z^{\{1\}}$ is also a solution, i.e. $S^{\{1\}} = \{z^{\{1\}} \oplus v : v \in \mathbb{B}^3\}$.

Similarly, for $W = \{2, 3\}$ we have the set $C^W = \{C_2, C_3\}$ which is minimal covering of I_0 with minimal solution $z^W = z^{\{2,3\}} = (0, .6, .6)$. Again, also vectors $z \geq z^{\{2,3\}}$ are solutions to the system, i.e. $S^{\{2,3\}} = \{z^{\{2,3\}} \oplus v : v \in \mathbb{B}^3\}$

Final solution set for the system (8) is then represented by the union of particular solution sets, $\bigcup_W S^W$.

3. Max-min eigenproblem

Let us first consider some two-dimensional examples that show how the solution to the max-min eigenproblem (1) splits into subsets of (K, L) -eigenvectors.

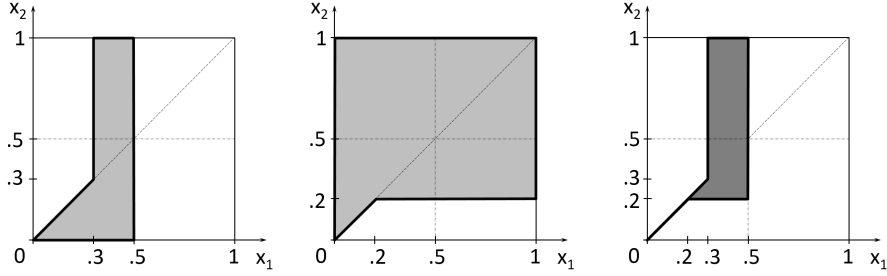


Figure 1: Sets X' , X'' and their intersection: the 0.5-eigenspace of A

Example 3.1. Take

$$A = \begin{pmatrix} .7 & .3 \\ .2 & .5 \end{pmatrix} \quad (15)$$

and consider $\lambda = .5$. Then the solution of (1) is equivalent to the system

$$\max(\min(.7, x_1), \min(.3, x_2)) = \min(.5, x_1), \quad (16)$$

$$\max(\min(.2, x_1), \min(.5, x_2)) = \min(.5, x_2) \quad (17)$$

The solution set for (16) is

$$X' = \{(x_1, x_2): ((x_1 \geq .3) \vee (x_1 \geq x_2)) \wedge (x_1 \leq .5)\}, \quad (18)$$

and the solution set for 17 is

$$X'' = \{(x_1, x_2): (x_2 \geq .2) \vee (x_1 \leq x_2)\}. \quad (19)$$

Then, the solution set to the eigenproblem is $X' \cap X''$. Sets X' , X'' and their intersection are displayed in Figure 1.

In Figure 2 we observe the effect of the value λ on the final solution set. The eigenvectors can be thus studied in individual areas (subsets) defined by λ . The boundaries of these areas represented by the dashed line divide the solution set of our two-dimensional example into four areas (in figure quadrants $Q_1 - Q_4$).

For the eigenvectors in Q_1 it holds that all $x_i \geq \lambda$ and thus we say that all $i \in L$ and $K = \emptyset$. We call these eigenvectors the background eigenvectors of A . For Q_2 and Q_3 it holds that $x_i \leq \lambda$ for some $i \in K$ and some $x_i \geq \lambda$ for some $i \in L$. In Q_4 all $x_i \leq \lambda$, it means that all $i \in K$ and $L = \emptyset$. We call these vectors the principal eigenvectors of A .

Note that in example 3.1 we have some “genuine” (K, L) -eigenvectors in the interior of Q_2 , which are neither principal nor background eigenvectors.

Example 3.2. In this example, we take

$$B = \begin{pmatrix} .4 & .5 \\ .2 & .5 \end{pmatrix} \quad (20)$$

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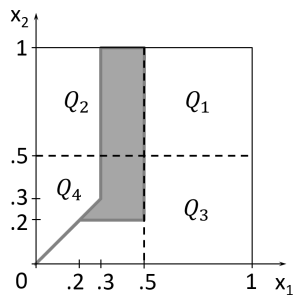


Figure 2: (K, L) -eigenvectors

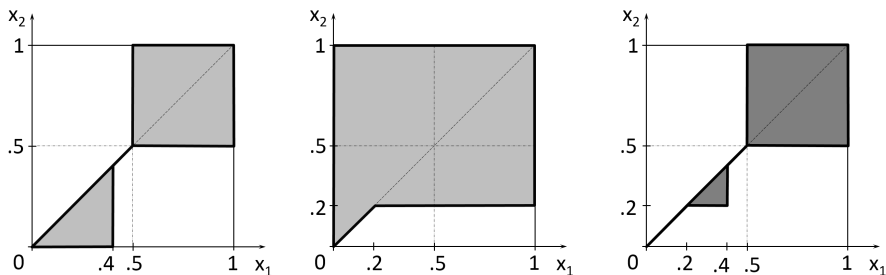


Figure 3: Sets Y' and Y'' and their intersection: the 0.5-eigenspace of B

and consider the same $\lambda = .5$. To solve (1) for $A = B$ we need to solve the system

$$\max(\min(.4, x_1), \min(.5, x_2)) = \min(.5, x_1), \tag{21}$$

$$\max(\min(.2, x_1), \min(.5, x_2)) = \min(.5, x_2) \tag{22}$$

Note that (22) is the same as (17). The solution set for (21) is

$$Y' = \left\{ (x_1, x_2) : ((x_1 = x_2 \leq .5) \vee (x_2 \leq x_1 \leq .4) \vee ((x_1 \geq .5) \wedge (x_2 \geq .5))) \right\}, \tag{23}$$

and the solution set for (22) is $Y'' = X''$ expressed in (19). Solution sets Y' and Y'' and their intersection are depicted in Figure 3 (there are no “genuine” (K, L) -eigenvectors in this example).

We are now going to give a theoretical description of background eigenvectors, principal eigenvectors and (K, L) -eigenvectors in max-min algebra.

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3.1. Background λ -eigenvectors

These are the vectors that satisfy $Ax = \lambda \mathbf{1}$ and $x_i \geq \lambda$ for all i . Let us introduce the following notation:

$$N^{>\lambda} = \{k: \max_{1 \leq i \leq n} a_{ik} > \lambda\}, \quad N^{\leq\lambda} = \{k: \max_{1 \leq i \leq n} a_{ik} \leq \lambda\}. \quad (24)$$

The set of background eigenvectors can be described as follows.

Proposition 3.1. *Let $A \in \mathbb{B}^{n \times n}$ and $\lambda \in \mathbb{B}$. Then the set of background λ -eigenvectors of A is nonempty if and only if*

$$\max_{1 \leq j \leq n} a_{ij} \geq \lambda \quad \forall i. \quad (25)$$

If (25) holds then the set of background eigenvectors is given by

$$\{x: x_k = \lambda \text{ for } k \in N^{>\lambda}, \lambda \leq x_k \leq 1 \text{ for } k \in N^{\leq\lambda}\}. \quad (26)$$

PROOF: Observe first that if there exist i with $a_{ij} < \lambda$ for all j , then also $a_{ij}x_j < \lambda$ for all j implying that $\bigoplus_{j=1}^n a_{ij}x_j = \lambda$ cannot hold and the set of background eigenvectors is empty. If (25) holds then the constant vector $x = \lambda \mathbf{1}$ satisfies $Ax = \lambda \mathbf{1}$ hence the set of background eigenvectors is nonempty.

If $k \in N^{>\lambda}$ then there exists i that $a_{ik} > \lambda$, and we need $x_k = \lambda$ to make sure that $\bigoplus_{j=1}^n a_{ij}x_j \leq \lambda$. This shows that the set of background λ -eigenvectors is a subset of (26).

Now take a vector from (26). Obviously, it satisfies $x_i \geq \lambda$ for each i . Since also $\max_j a_{ij} \geq \lambda$ for all j , we have $\bigoplus_j a_{ij}x_j \geq \lambda$. But we also have $\bigoplus_j a_{ij}x_j \leq \lambda$. Indeed, since $x_j = \lambda$ in (26) whenever $j \in N^{>\lambda}$, that is, whenever there exists i with $a_{ij} > \lambda$, we have $a_{ij}x_j \leq \lambda$ for all such j and all i . For $j \in N^{\leq\lambda}$, inequality $a_{ij}x_j \leq \lambda$ follows from $\max_i a_{ij} \leq \lambda$ (by the definition of $N^{\leq\lambda}$).

The proof is complete. \square

Note that the set of background eigenvectors is a max-min convex set, and not a max-min space. The proof of the following corollary of Proposition 3.2 is straightforward and will be omitted.

Corollary 3.1. *If the set of background eigenvectors is non-empty, then it is the max-min convex hull of $e^{\lambda 0} = \lambda \mathbf{1}$ and vectors $e^{\lambda k}$ for $k \in N^{\leq\lambda}$, for which*

$$e_i^{\lambda k} = \begin{cases} 1, & \text{if } k = i, \\ \lambda, & \text{otherwise.} \end{cases}$$

The following proposition will be helpful when describing more general sets of the form (26) as max-min convex hulls.

Proposition 3.2. *Suppose that*

$$S = \{x: x_k = \lambda \text{ for } k \in N_1, \lambda \leq x_k \leq 1 \text{ for } k \in N_2\}, \quad (27)$$

where N_1 and N_2 are such that $N_1 \cup N_2 = N (= \{1, \dots, n\})$ and $N_1 \cap N_2 = \emptyset$. Then

$$S = \{\lambda \mathbf{1} \oplus z_{N_2}: z_{N_2} \in \mathbb{B}^{|N_2|}\} \quad (28)$$

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3.2. Principal λ -eigenvectors

These are the vectors that satisfy $Ax = x$ and $x_i \leq \lambda$ for all i . Description of a generating set of the space of principal max-min λ -eigenvectors is given below. Observe that any principal λ -eigenvector is a principal eigenvector and therefore we can apply Theorem 2.1.

Corollary 3.2 (Gondran-Minoux [18], Ch. 6, Corollary 3.5). *The set of principal λ -eigenvectors of $A \in \mathbb{B}^{n \times n}$ is the max-min column space $\{A_\lambda^* y : y \in \mathbb{B}^n\}$, where the columns of A_λ^* are defined by*

$$(A_\lambda^*)_{\cdot i} = \lambda a_{ii}^+(A^*)_{\cdot i} : i = 1, \dots, n. \quad (29)$$

More precisely, each vector of (29) is a principal λ -eigenvector, and each principal λ -eigenvector x can be represented as

$$x = \bigoplus_{i \in C(A, x)} \lambda x_i a_{ii}^+(A^*)_{\cdot i}, \quad (30)$$

where $C(A, x) \subseteq N$ is a set containing a node from each strongly connected component of $\text{Sat}(A, x)$.

PROOF: The claim follows as an easy corollary of Theorem 2.1. Indeed, since $a_{ii}^+(A^*)_{\cdot i}$ satisfies $Ax = x$, so does $\lambda a_{ii}^+(A^*)_{\cdot i}$. As components of this vector do not exceed λ , it is a principal λ -eigenvector. Letting x be a principal λ -eigenvector, we see that it satisfies (5) since it satisfies $Ax = x$. Equation (30) follows from (5) after multiplying both parts of (5) by λ and observing that $\lambda x = x$ since $x_i \leq \lambda$ for all i . \square

3.3. (K, L) λ -eigenvectors

Now we consider (K, L) max-min λ -eigenvectors, i.e., $x \in \mathbb{B}^n$ such that $Ax = \lambda x$, $x_i \leq \lambda$ for $i \in K$ and $x_i \geq \lambda$ for $i \in L$, where $K, L \subseteq \{1, \dots, n\}$ are such that $K \cup L = \{1, \dots, n\}$ and $K \cap L = \emptyset$.

By this definition, every (K, L) λ -eigenvector is of the form $x = (x_L \ x_K)$ (in a suitable permutation of indices) with

$$A_{KK}x_K \oplus A_{KL}x_L = x_K, \quad x_K \leq \lambda \mathbf{1}_K \quad (31)$$

$$A_{LK}x_K \oplus A_{LL}x_L = \lambda \mathbf{1}_L, \quad x_L \geq \lambda \mathbf{1}_L \quad (32)$$

We start with describing the solvability and the set of solutions for (31).

Proposition 3.3. *Equation (31) is solvable (together with $x_K \leq \lambda \mathbf{1}_K$) with respect to x_K if and only if*

$$(A_{KK})^* A_{KL} x_L \leq \lambda \mathbf{1}_K, \quad (33)$$

and then the set of its solutions for fixed x_L is given by

$$\begin{aligned} S_K(x_L, \lambda) &= \{(A_{KK})^* A_{KL} x_L \oplus v : A_{KK} v = v, v \leq \lambda \mathbf{1}_K\} \\ &= \{(A_{KK})^* A_{KL} x_L \oplus (A_{KK})_\lambda^* z_K : z_K \in \mathbb{B}^{|K|}\} \end{aligned} \quad (34)$$

where $(A_{KK})_\lambda^*$ is defined as in (29).

PROOF: the assertion follows from Theorem 2.2. \square

Before describing the solvability of (32), denote

$$L_1 = \{i \in L: \bigoplus_{j \in L} a_{ij} \geq \lambda\}, \quad L_2 = \{i \in L: \bigoplus_{j \in L} a_{ij} < \lambda\} \quad (35)$$

Proposition 3.4. *Equation (32), considered together with $x_K \leq \lambda \mathbf{1}_K$, is equivalent to*

$$A_{L_1 L} x_L = \lambda \mathbf{1}_{L_1}, \quad x_L \geq \lambda \mathbf{1}_L, \quad (36)$$

$$A_{L_2 K} x_K = \lambda \mathbf{1}_{L_2}, \quad x_K \leq \lambda \mathbf{1}_K. \quad (37)$$

PROOF: In terms of the notation (35), equation (32) is written as follows:

$$A_{L_1 K} x_K \oplus A_{L_1 L} x_L = \lambda \mathbf{1}_{L_1}, \quad (38)$$

$$A_{L_2 K} x_K \oplus A_{L_2 L} x_L = \lambda \mathbf{1}_{L_2}, \quad (39)$$

$$x_L \geq \lambda \mathbf{1}_L, \quad x_K \leq \lambda \mathbf{1}_K. \quad (40)$$

We now show that (38), (39), (40) are equivalent to (36) and (37). Indeed, $x_L \geq \lambda \mathbf{1}_L$ together with $\bigoplus_{j \in L} a_{ij} \geq \lambda$ for $i \in L_1$ imply that $A_{L_1 L} x_L \geq \lambda \mathbf{1}_{L_1}$ holds for any feasible x , and this makes the first term in (38) redundant, also since $x_K \leq \lambda \mathbf{1}_K$ implies $A_{L_1 K} x_K \leq \lambda \mathbf{1}_{L_1}$. As for the second term in (39), we have $\bigoplus_{j \in L} a_{ij} < \lambda$ for $i \in L_2$ implying that $A_{L_2 L} x_L < \lambda \mathbf{1}_{L_2}$ and making this term redundant as well. \square

To solve (36) let us introduce the following notation in analogy with (24):

$$L^{>\lambda, L_1} = \{k \in L: \max_{i \in L_1} a_{ik} > \lambda\} \quad (41)$$

$$L^{>\lambda, K} = \{\ell \in L: \max_{k \in K} ((A_{KK})^* A_{KL})_{k\ell} > \lambda\} \quad (42)$$

$$L' = L^{>\lambda, L_1} \cup L^{>\lambda, K}, \quad \tilde{L} = L \setminus L'. \quad (43)$$

Proposition 3.5. *The solution set to (36) with condition (33) is given by*

$$S_L(\lambda) = \{\lambda \mathbf{1}_L \oplus z_{\tilde{L}}: z_{\tilde{L}} \in \mathbb{B}^{|\tilde{L}|}\}. \quad (44)$$

PROOF: Following Proposition 3.1, the set of solutions to (36) is

$$\{x_L: x_\ell = \lambda \text{ for } \ell \in L^{>\lambda, L_1}, \lambda \leq x_\ell \leq 1 \text{ for } \ell \notin L^{>\lambda, L_1}\}. \quad (45)$$

However, we also have (33), which is satisfied whenever

$$x_\ell = \lambda \text{ for all } \ell \in L^{>\lambda, K}. \quad (46)$$

The set of solutions to (36) with respect to both conditions (45) and (46) can be written as

$$S_L(\lambda) = \{x_L: x_\ell = \lambda \text{ for } \ell \in L', \lambda \leq x_\ell \leq 1 \text{ for } \ell \in \tilde{L}\}. \quad (47)$$

Using Proposition 3.2 we can express $S_L(\lambda)$ as shown in (44). \square

Proposition 3.6. *The set of vectors x_K solving (31) for a given $x_L = \lambda \mathbf{1}_L \oplus z_{\bar{L}}$ is of the form*

$$S_K(x_L, \lambda) = \{\lambda(A_{KK})^* A_{KL} \mathbf{1}_L \oplus (A_{KK})^* A_{K\bar{L}} z_{\bar{L}} \oplus (A_{KK})^*_{\lambda} z_K : z_K \in \mathbb{B}^{|K|}\}. \quad (48)$$

PROOF: The computation of $S_K(x_L, \lambda)$ in view of (34) uses $x_L \in S_L(\lambda)$ according to (44) and distributivity.

$$\begin{aligned} S_K(x_L, \lambda) &= \{(A_{KK})^* A_{KL} x_L \oplus (A_{KK})^*_{\lambda} z_K : z_K \in \mathbb{B}^{|K|}\} \\ &= \{\lambda(A_{KK})^* A_{KL} \mathbf{1}_L \oplus (A_{KK})^* A_{K\bar{L}} z_{\bar{L}} \oplus (A_{KK})^*_{\lambda} z_K : z_K \in \mathbb{B}^{|K|}\}. \end{aligned} \quad (49)$$

□

Substituting $x_K \in S_K(x_L, \lambda)$ given by (48) into (37), we obtain the following equation:

$$\lambda A_{L_2 K} (A_{KK})^* A_{KL} \mathbf{1}_L \oplus A_{L_2 K} (A_{KK})^* A_{K\bar{L}} z_{\bar{L}} \oplus A_{L_2 K} (A_{KK})^*_{\lambda} z_K = \lambda \mathbf{1}_{L_2} \quad (50)$$

It can be seen that the system (50) is of the form $A' z' \oplus b' = \lambda \mathbf{1}$ where

$$\begin{aligned} A' &= (A_{L_2 K} (A_{KK})^* A_{K\bar{L}}, A_{L_2 K} (A_{KK})^*_{\lambda}), \\ b' &= \lambda A_{L_2 K} (A_{KK})^* A_{KL} \mathbf{1}_L, \quad z' = (z_{\bar{L}} \ z_K)^T. \end{aligned} \quad (51)$$

This observation enables us to describe the solution $z' = (z_{\bar{L}} \ z_K)^T$ using the results of Subsection 2.2, as below.

Proposition 3.7. *The minimal solutions $(z')^W = (z_K^W z_{\bar{L}}^W)$ of $A' z' \oplus b' = \lambda \mathbf{1}$ with A' and b' as in (51) are defined by*

$$(z')_j^W = \begin{cases} \lambda, & \text{if } j \in W, \\ 0, & \text{otherwise.} \end{cases} \quad (52)$$

where $\{\cup C_j : j \in W\}$ with

$$C_j = \{i \in I_0 : (A')_{ij} = \lambda\} \quad \text{for } j \in \tilde{N}, \quad (53)$$

is a minimal covering of

$$I_0 = \{i \in L_2 : A_{iK} (A_{KK})^* A_{KL} \mathbf{1}_L < \lambda\}. \quad (54)$$

Proposition 3.8. *The set of all solutions of $A' z' \oplus b' = \lambda \mathbf{1}$ with A' and b' as in (51) is given by $\cup_W S_W$ where the union is taken over all minimal coverings $\{\cup C_j : j \in W\}$ of I_0 and $S_W = \{z^W \oplus v_{\tilde{N}} : v_{\tilde{N}} \in \mathbb{B}^{|\tilde{N}|}\}$*

PROOF:[Proof of Proposition 3.7 and 3.8] Observe that all entries of $A_{KK}^* A_{K\bar{L}}$ do not exceed λ by (42) and the definition of \tilde{L} , and that all coefficients of

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$(A_{KK})_\lambda^*$ do not exceed λ by (29). Hence the entries of A' and b' do not exceed λ , and all solutions of $A'z' \oplus b' = \lambda \mathbf{1}$ can be found as in Corollary 2.1 and using (14), with A' and b' instead of A and b , $\tilde{N} = \tilde{L} \cup K$ instead of N . \square

The next theorem, which describes the set of (K, L) λ -eigenvectors, is the main result of the section. This theorem follows from the arguments written above, but we also include a formal proof based on backtracking the above arguments.

Theorem 3.1. *Let $A \in \mathbb{B}^{n \times n}$, $\lambda \in \mathbb{B}$ and K, L such that $K \cup L = N$ and $K \cap L = \emptyset$ be given. Then (K, L) -eigenvector exists if and only if (50) is solvable, which happens if and only if $I_0 = \cup_{j \in \tilde{N}} C_j$, with I_0 and C_j defined as in (53) and (54) and $\tilde{N} = \tilde{L} \cup K$.*

In this case, $x = (x_K \ x_L)^T$ is a (K, L) -eigenvector of A with eigenvalue λ if and only if x_K and x_L can be expressed as follows:

$$\begin{aligned} x_L &= \lambda \mathbf{1}_L \oplus v_{\tilde{L}}, \\ x_K &= \lambda (A_{KK})^* A_{KL} \mathbf{1}_L \oplus (A_{KK})^* A_{K\tilde{L}} v_{\tilde{L}} \oplus (A_{KK})_\lambda^* (z_K^W \oplus v_K), \end{aligned} \quad (55)$$

where $v_K \in \mathbb{B}^{|K|}$, $v_{\tilde{L}} \in \mathbb{B}^{|\tilde{L}|}$, and z^W is the minimal solution of (50) corresponding to a minimal covering $\{C_j : j \in W\}$ of I_0 .

PROOF: Consider a set of vectors $z_{\tilde{N}} = (z_K \ z_{\tilde{L}})$ described by

$$z_K = z_K^W \oplus v_K, \quad z_{\tilde{L}} = z_{\tilde{L}}^W \oplus v_{\tilde{L}}, \quad v_K \in \mathbb{B}^{|K|}, \quad v_{\tilde{L}} \in \mathbb{B}^{|\tilde{L}|}$$

for a minimal W such that $\cup_{j \in W} C_j = I_0$. By Proposition 3.7 and Proposition 3.8, the union of these sets over W comprises the set of all solutions to (50), which is solvable if and only if $\cup_{j \in \tilde{N}} C_j = I_0$.

Equation (50), if it is solvable, is the same as $A_{L_2 K} x_K = \lambda \mathbf{1}_{L_2}$ of (37), where $x_K \in S_K(x_L, \lambda)$ with the latter set described in (48). Note that the condition $x_K \leq \lambda \mathbf{1}_K$ immediately follows from (48). Then we see that (55) describes all vectors $(x_K \ x_L)$ such that $x_L \in S_L(\lambda)$, $x_K \in S_K(x_L, \lambda)$ and condition (37) is satisfied, after observing that $x_L = \lambda \mathbf{1}_L \oplus z_{\tilde{L}}^W \oplus v_{\tilde{L}} = \lambda \mathbf{1} \oplus v_{\tilde{L}}$ (since each component of z^W is less than or equal to λ).

By Proposition 3.6, $S_K(x_L, \lambda)$ is the set of all vectors x_K solving (31) for a given $x_L = \lambda \mathbf{1}_L \oplus z_{\tilde{L}}^W$, which is by Proposition 3.5 a general solution to (36) with condition (33). Observe that by Proposition 3.3 condition (33) is equivalent to solvability of (31) with respect to x_K for a given x_L . Thus (55) describes all vectors $(x_K \ x_L)$ that satisfy (31), (36) and (37) simultaneously.

Furthermore, in view of Proposition 3.4, (36) and (37) can be replaced by (32), implying that (55) yields all solutions to (31) and (32) if (50) is solvable.

If (50) is not solvable then (37) is not compatible with (31) and (36). The proof is complete. \square

Let us also note the following special case of the above considerations.

Corollary 3.3. *If $L_2 = \emptyset$ then the solution set to (31), (32) is the set of $x = (x_L \ x_K)$ given by*

$$\begin{aligned} x_L &= \lambda \mathbf{1}_L \oplus z_{\bar{L}}, \\ x_K &= \lambda (A_{KK})^* A_{KL} \mathbf{1}_L \oplus (A_{KK})^* A_{K\bar{L}} z_{\bar{L}} \oplus (A_{KK})^*_{\lambda} z_K, \end{aligned} \quad (56)$$

where parameters $z_{\bar{L}} \in \mathbb{B}^{|\bar{L}|}$, $z_K \in \mathbb{B}^{|K|}$ have arbitrary values.

In this case (50) is trivially solvable, as an “empty equation”.

Theorem 3.1 gives us a clear way to generate all (K, L) -eigenvectors. If we want to “test” its validity, we may recall that, since it can be easily shown that $Ax \leq x$ and $A^*x = x$ are equivalent [18] and since any λ -eigenvector satisfies $Ax = \lambda x \leq x$, we should have $A^*x = x$ for any λ -eigenvector x . Let us show that any vector given by (55) satisfies this property.

Corollary 3.4. *Let $x = (x_K \ x_L)$ satisfy (55). Then we have $A^*x = x$.*

PROOF: Since $A^*x = x$ is equivalent to $Ax \leq x$, we will prove the latter. Inequality $Ax \leq x$ can be written as the following four:

$$A_{KK}x_K \leq x_K, \quad A_{KL}x_L \leq x_K, \quad A_{LK}x_K \leq x_L, \quad A_{LL}x_L \leq x_L.$$

The inequality $A_{KK}x_K \leq x_K$ follows since x_K , by the second equation of (55), satisfies $(A_{KK})^*v = x_K$ for some v and we have $A_{KK}(A_{KK})^* = (A_{KK})^+ \leq (A_{KK})^*$. As for the inequality $A_{KL}x_L \leq x_K$, we have

$$A_{KL}x_L = \lambda A_{KL} \mathbf{1}_L \oplus A_{K\bar{L}} v_{\bar{L}} \leq x_K.$$

As for the inequality $A_{LK}x_K \leq x_L$, it follows since x_L and x_K of (55) satisfy $x_L \geq \lambda \mathbf{1}$ and $x_K \leq \lambda \mathbf{1}$.

It remains to prove the inequality $A_{LL}x_L \leq x_L$, which we are going to do for arbitrary x_L belonging to the set defined in (45), using that this set contains (44) and hence any x_L given by the first equation of (55). Denoting $L_3 = L^{>\lambda, L_1}$ and $L_4 = L \setminus L^{>\lambda, L_1}$ and seeing that $x_{L_3} = \lambda \mathbf{1}_{L_3}$ and $x_{L_4} \geq \lambda \mathbf{1}_{L_4}$, we have to prove

$$\lambda A_{L_3 L_3} \mathbf{1}_{L_3} \leq \lambda \mathbf{1}_{L_3}, \quad A_{L_3 L_4} x_{L_4} \leq \lambda \mathbf{1}_{L_3}, \quad \lambda A_{L_4 L_3} \mathbf{1}_{L_3} \leq x_{L_4}, \quad A_{L_4 L_4} x_{L_4} \leq x_{L_4}.$$

The first and the third of these inequalities are obvious. As for the inequalities $A_{L_3 L_4} x_{L_4} \leq \lambda \mathbf{1}_{L_3}$ and $A_{L_4 L_4} x_{L_4} \leq x_{L_4}$, we observe using the definition of $L^{>\lambda, L_1}$ (41) and L_2 in (35) that all entries in A_{LL_4} do not exceed λ , and recall that $x_{L_4} \geq \lambda \mathbf{1}_{L_4}$. Thus the inequality $A_{LL}x_L \leq x_L$ also follows and the proof is complete. \square

Let us now describe the set of (K, L) -eigenvectors as a max-min convex hull of a finite number of points.

Corollary 3.5. *If the set of (K, L) -eigenvectors associated with eigenvalue λ is non-empty, then it is the max-min convex hull of the points $x^{W,0} = (x_K^{W,0} \ x_L^{W,0})$, $x^{W,i} = (x_K^{W,i} \ x_L^{W,i})$ for $i \in \tilde{L}$ and $x^{W,k} = (x_K^{W,k} \ x_L^{W,k})$ for $k \in K$, where*

$$\begin{aligned} x_L^{W,0} &= \lambda \mathbf{1}_L, & x_K^{W,0} &= (A_{KK})^* A_{KL} x_L^{W,0} \oplus (A_{KK})_\lambda^* z_K^W, \\ x_L^{W,i} &= \lambda \mathbf{1}_L \oplus e_i, & x_K^{W,i} &= (A_{KK})^* A_{KL} x_L^{W,i} \oplus (A_{KK})_\lambda^* z_K^W, \\ x_L^{W,k} &= \lambda \mathbf{1}_L, & x_K^{W,k} &= (A_{KK})^* A_{KL} x_L^{W,k} \oplus (A_{KK})_\lambda^* (z_K^W \oplus e_k), \end{aligned} \quad (57)$$

and W ranges over all minimal coverings $\{C_j: j \in W\}$ of I_0 .

PROOF: Using that $\mathbb{B}^{|\tilde{N}|} = \mathbb{B}^{|K|} \times \mathbb{B}^{|\tilde{L}|}$ is a max-min convex set generated by the zero vector $\mathbf{0}$ and unit vectors e_i for $i \in \tilde{L}$ and e_k for $k \in K$, we substitute these generators for $v_{\tilde{N}} = (v_{\tilde{L}} \ v_K)$ in (55) and obtain that the set of (K, L) -eigenvectors is the max-min convex hull of the points described by (57). \square

For the purposes of computation note that the minimal solutions of (50), which correspond to minimal coverings of I_0 (54) by unions of C_j (53), can be found using the methods described in Elbassioni [9]. The set of all (K, L) -eigenvectors can be then efficiently described using Theorem 3.1 and Corollary 3.5.

In particular, the number of minimal solutions of $A'z' \oplus b' = \lambda \mathbf{1}$ is equal to the number of minimal coverings $\{\cup C_j: j \in W\}$ of I_0 , which depends on entries of A (particularly $(A_{KK})^*$) and given λ . However, the dependency of the points generating the set of (K, L) -eigenvectors as their max-min convex hull on these minimal coverings seems rather uncertain and can be eliminated if $(A_{KK})_\lambda^* z_K^W$ is dominated by the sum of other terms in (55).

In the following example, there is just one possible minimal covering that produces the only minimal solution used in further computation.

Example 3.3. The following three-dimensional example shows how to find the eigenspace for given matrix A and eigenvalue λ .

Take

$$A = \begin{pmatrix} .1 & .5 & .7 \\ 0 & .4 & .8 \\ .1 & .1 & .5 \end{pmatrix}, \quad \lambda = .5$$

First, let us introduce the notation of the vector with interval entries. The vector where entries are of form $a \leq x_1 \leq b$ and $c \leq x_2 \leq d$ is denoted in further text as

$$x = \begin{pmatrix} \langle a, b \rangle \\ \langle c, d \rangle \end{pmatrix}.$$

As the λ -eigenspace is a union of background, principal and (K, L) λ -eigenvectors, we are going to compute the solution for each individual case.

For the case of background λ -eigenvectors, first, we have to verify the existence of this eigenvector-type. From (25) we see that this set is nonempty. According to (26), background eigenvectors are all vectors of form

$$x = \begin{pmatrix} \langle .5, 1 \rangle \\ \langle .5, 1 \rangle \\ .5 \end{pmatrix} \quad (58)$$

Their set is the max-min convex hull of $(.5, .5, .5)$, $(1, .5, .5)$, $(.5, 1, .5)$.

Principal λ -eigenvectors are computed using Corollary 3.2. We find that generators of max-min eigenspace for $x_i \leq \lambda$ for all i are $(.1, .1, .1)$, $(.4, .4, .1)$ and $(.5, .5, .5)$. Of these vectors, the first one is obviously redundant as $(.1, .1, .1) = .1 \otimes (.5, .5, .5)$. If we want to describe this eigenspace as a max-min convex hull then the zero vector $(0, 0, 0)$ is routinely added to the generating set.

When computing (K, L) eigenvectors, first we have to determine particular (K, L) partition. For partition $K = \{1, 3\}$, $L = \{2\}$ we have

$$A_{KK} = \begin{pmatrix} .1 & .7 \\ .1 & .5 \end{pmatrix}, A_{KL} = \begin{pmatrix} .5 \\ .1 \end{pmatrix}, A_{LK} = (0 \quad .8), A_{LL} = (.4),$$

$$(A_{KK})^* = \begin{pmatrix} 1 & .7 \\ .1 & 1 \end{pmatrix}, (A_{KK})_\lambda^* = \begin{pmatrix} .1 & .5 \\ .1 & .5 \end{pmatrix}.$$

We are solving the system

$$\begin{pmatrix} .1 & .7 \\ .1 & .5 \end{pmatrix} \otimes \begin{pmatrix} x_1 \\ x_3 \end{pmatrix} \oplus \begin{pmatrix} .5 \\ .1 \end{pmatrix} \otimes x_2 = \begin{pmatrix} x_1 \\ x_3 \end{pmatrix} \quad (59)$$

$$(0 \quad .8) \otimes \begin{pmatrix} x_1 \\ x_3 \end{pmatrix} \oplus (.4) \otimes x_2 = \lambda \quad (60)$$

By verifying (33), we find out that (33) holds for all values x_2 , and thus a solution to (59) exists.

We can also express the sets $L_1 = \emptyset$, $L_2 = \{2\}$, $L' = \emptyset$ and $\tilde{L} = \{2\}$. As $L_2 \neq \emptyset$, we need to find the solution set to (50):

$$\begin{aligned} &.5 \otimes (0 \quad .8) \otimes \begin{pmatrix} 1 & .7 \\ .1 & 1 \end{pmatrix} \otimes \begin{pmatrix} .5 \\ .1 \end{pmatrix} \otimes 1 \oplus \\ &(0 \quad .8) \otimes \begin{pmatrix} 1 & .7 \\ .1 & 1 \end{pmatrix} \otimes \begin{pmatrix} .5 \\ .1 \end{pmatrix} \otimes 1 \otimes z_2 \oplus (0 \quad .8) \otimes \begin{pmatrix} .1 & .5 \\ .1 & .5 \end{pmatrix} \otimes \begin{pmatrix} z_1 \\ z_3 \end{pmatrix} = .5, \end{aligned} \quad (61)$$

which is the same as

$$.1 \oplus (.1 \quad .1 \quad .5) \otimes \begin{pmatrix} z_1 \\ z_2 \\ z_3 \end{pmatrix} = .5 \quad (62)$$

The only minimal (and hence the least) solution to this system is $z^{\{3\}} = (0, 0, .5)$, so the solution set is $\{z: z^{\{3\}} \leq z \leq 1\}$. We can also express $z_L^{\{3\}} = z_L^{\{3\}} = (0)$, $z_K^{\{3\}} = (0, .5)$.

Following Theorem 3.1 and using equations (55) we obtain

$$\begin{aligned}
x_2 &= .5 \oplus v_2, \\
\begin{pmatrix} x_1 \\ x_3 \end{pmatrix} &= .5 \begin{pmatrix} 1 & .7 \\ .1 & 1 \end{pmatrix} \begin{pmatrix} .5 \\ .1 \end{pmatrix} \oplus \begin{pmatrix} 1 & .7 \\ .1 & 1 \end{pmatrix} \begin{pmatrix} .5 \\ .1 \end{pmatrix} v_2 \\
&\oplus \begin{pmatrix} .1 & .5 \\ .1 & .5 \end{pmatrix} \left(\begin{pmatrix} 0 \\ .5 \end{pmatrix} \oplus \begin{pmatrix} v_1 \\ v_3 \end{pmatrix} \right) \\
&= \begin{pmatrix} .5 \\ .1 \end{pmatrix} (1 \oplus v_2) \oplus \begin{pmatrix} .5 \\ .5 \end{pmatrix} = \begin{pmatrix} .5 \\ .5 \end{pmatrix},
\end{aligned} \tag{63}$$

which can be written as

$$x = \begin{pmatrix} .5 \\ \langle .5, 1 \rangle \\ .5 \end{pmatrix} \tag{64}$$

For the partition $K = \{2, 3\}$, $L = \{1\}$ we are solving the system

$$\begin{pmatrix} .4 & .8 \\ .1 & .5 \end{pmatrix} \otimes \begin{pmatrix} x_2 \\ x_3 \end{pmatrix} \oplus \begin{pmatrix} 0 \\ .1 \end{pmatrix} \otimes x_1 = \begin{pmatrix} x_2 \\ x_3 \end{pmatrix} \tag{65}$$

$$(.5 \ .7) \otimes \begin{pmatrix} x_2 \\ x_3 \end{pmatrix} \oplus (.1) \otimes x_1 = \lambda \tag{66}$$

Similarly to previous procedures we compute eigenvectors for this partition

$$x = \begin{pmatrix} \langle .5, 1 \rangle \\ .5 \\ .5 \end{pmatrix} \tag{67}$$

Thus we see that (K, L) -eigenvectors for $L = \{1\}$ and $L = \{2\}$ are background eigenvectors. By similar routine calculations we can check that the same is true for (K, L) -eigenvectors for any non-empty L , and this is different from Example 3.1, where we have (K, L) -eigenvectors that are neither background nor principal. Note that in Example 3.3 any set of (K, L) -eigenvectors is non-empty, as $(.5, .5, .5)$ is a λ -eigenvector, which is shared between all of them.

The whole λ -eigenspace, shown on Figure 4, is thus the union of two parts: principal eigenvectors and background eigenvectors. This eigenspace (as a union of max-min convex hulls which is itself max-min convex) is the max-min convex hull of: $(.5, .5, .5)$, $(1, .5, .5)$, $(.5, 1, .5)$, $(.4, .4, .1)$ and $(0, 0, 0)$. Removing the zero vector we obtain a generating set of the whole λ -eigenspace, and next we notice that $(.5, .5, .5)$ is redundant, since for example $(.5, .5, .5) = .5 \otimes (1, .5, .5)$. Thus the λ -eigenspace is generated by $(1, .5, .5)$, $(.5, 1, .5)$ and $(.4, .4, .1)$.

4. Discussion

Theorem 3.1 and Corollary 3.5 provide us with a description of the sets of (K, L) -eigenvectors associated with eigenvalue λ and, in particular, give us a

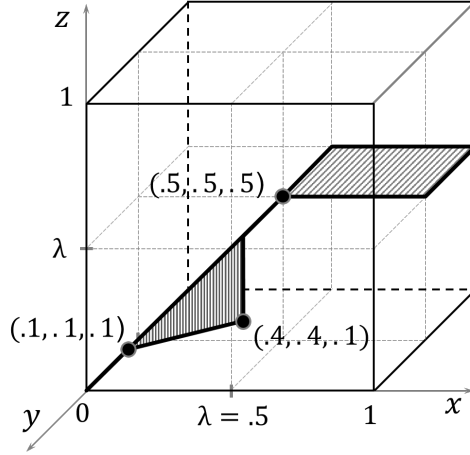


Figure 4: Eigenspace for A and λ of Example 3.3

finite set of points such that the set of (K, L) max-min eigenvectors appears as the max-min convex hull of them. Collecting such points for all sets of (K, L) -eigenvectors we obtain a finite set of generators for the whole λ -eigenspace.

There are some problems with this description. Clearly, there are exponentially many partitions $K \cup L = \{1, \dots, n\}$, and to implement the above described method for finding the generators of λ -eigenspace one should find a way to quickly eliminate many partitions that are redundant: in two of the three examples that we considered, only the trivial partitions, corresponding to principal and background eigenvectors, were important. The procedure also raises the question about the number of points that generate the λ -eigenspace. From Example 3.2, it is clear that the number of such points can exceed the dimension, as the eigenspace in this example is generated by $(.4, .2)$, $(.5, 1)$, $(1, .5)$ (e.g., consider it as a “max-min quadrangle” between these points and the zero vector, and use the forms of max-min segments given in [24]), with none of these points being redundant. However, it is not known how quickly the minimal number of such generators can grow with matrix dimension.

Similar questions can be asked also about the set of (K, L) -eigenvectors. The points that generate this set as their max-min convex hull come from certain minimal coverings, whose number may also grow quickly with matrix dimension, unless $L_2 = \emptyset$ as in Corollary 3.3. However, we have seen that the terms coming from these minimal coverings are often dominated by other terms, thus decreasing the number of points making up the max-min convex hull.

There is a perspective to develop an application of max-min λ -eigenspaces in medical diagnostics following the ideas of Sanchez [27]. Also, there are still many unresolved questions in the geometry over max-min semiring: about the generating sets of max-min eigenspaces and more general max-min linear spaces,

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9 and more generally, about max-min polytopes.

10 11 12 **5. Acknowledgements**

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15 16 17 **References**

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**On the new revision of “The theory of max-min (K, L) -
eigenvectors”.**

In the new revision we made the following notable changes, following the referee’s suggestions:

- We changed the title and the abstract following referee’s suggestions, although not literally, as we did not like putting ‘max-min eigenvectors’ and ‘fuzzy algebra’ in the same sentence. However, it is right to mention fuzzy algebra as a synonym of max-min algebra, and to add fuzzy algebra to keywords.
- We renamed pure λ -eigenvectors to principal λ -eigenvectors.
- We made all other minor corrections as suggested by the referee.

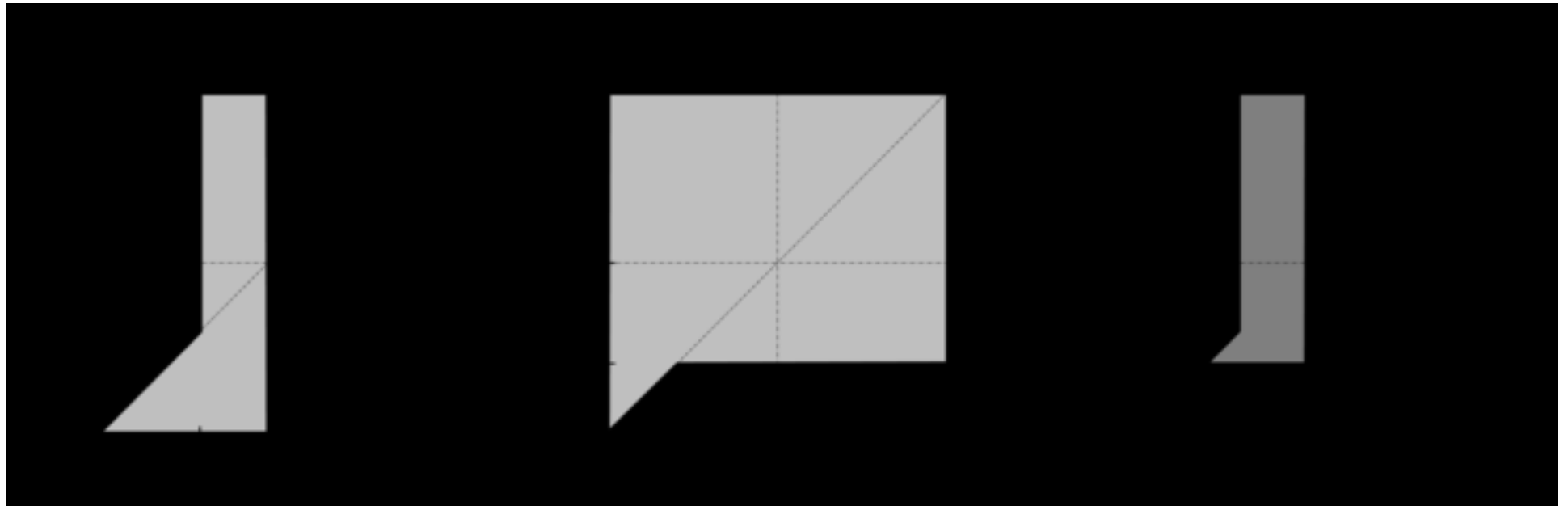
That said, our initial position on developing practical applications of our theory stays the same. The two examples provided in introduction are a mere indication that such an application can be developed. While we are grateful to the referee for challenging us to make some improvement also in this part, we maintain that developing a practical application of our theory is not a purpose of this paper. For this reason, we did not make any major development in our description of possible applications, but only added a minor explanation in Example 1.2, about the data security (just before $y = x$).

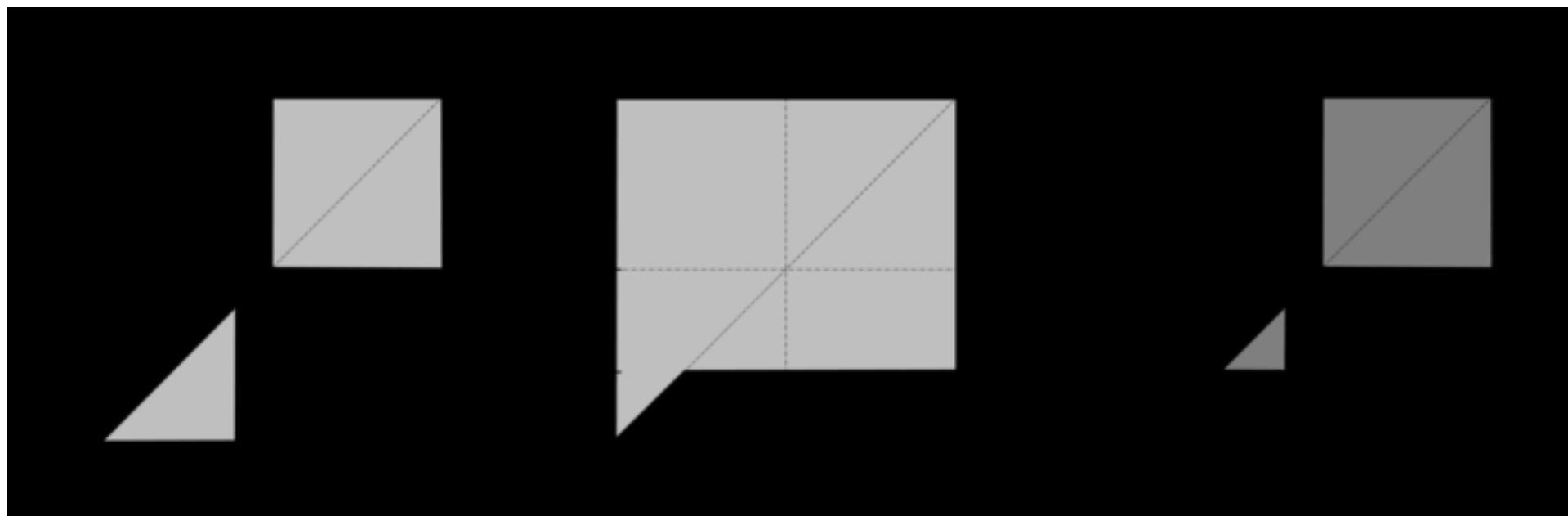
Best Wishes,
The authors.

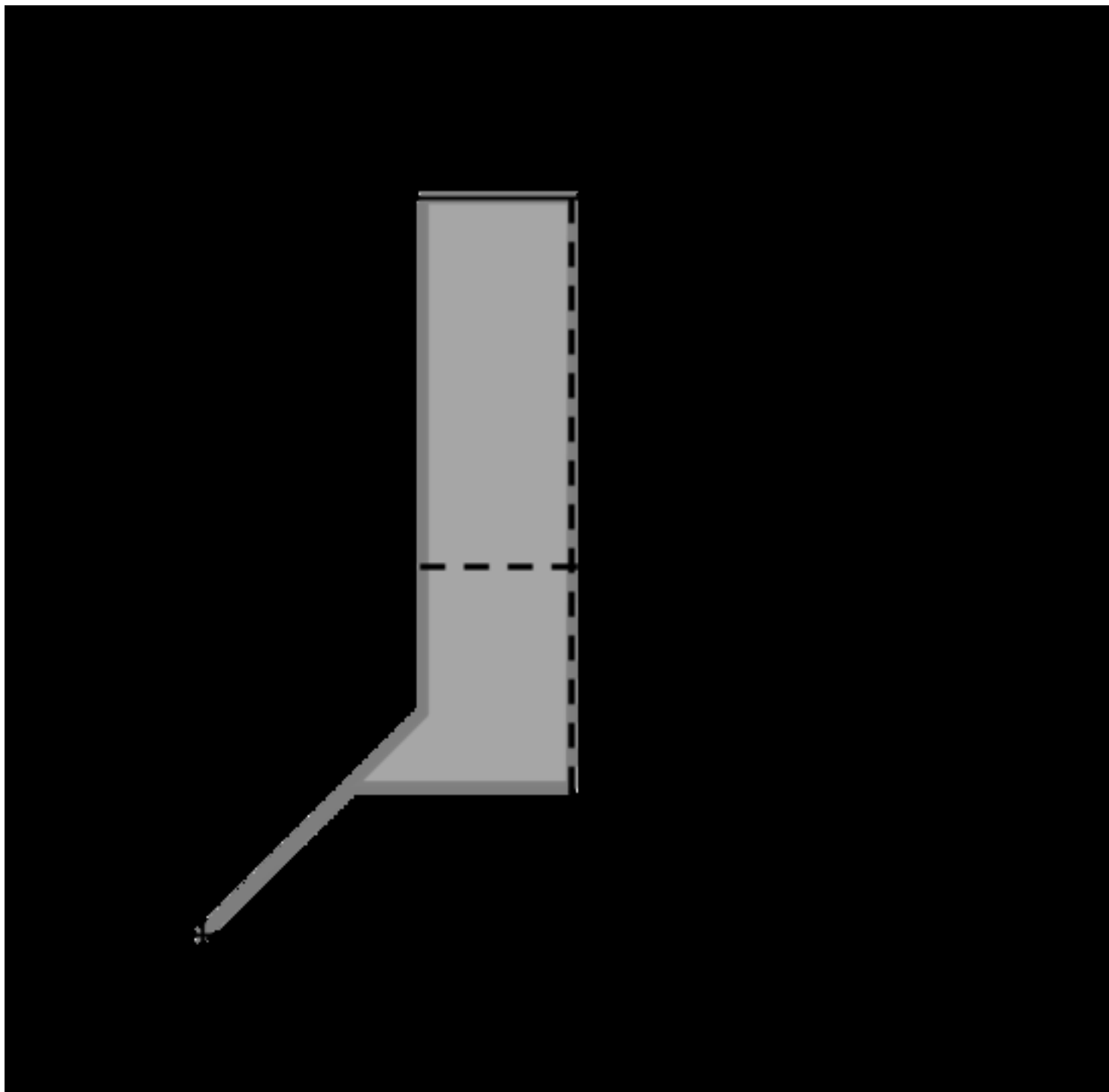
Declaration of interests

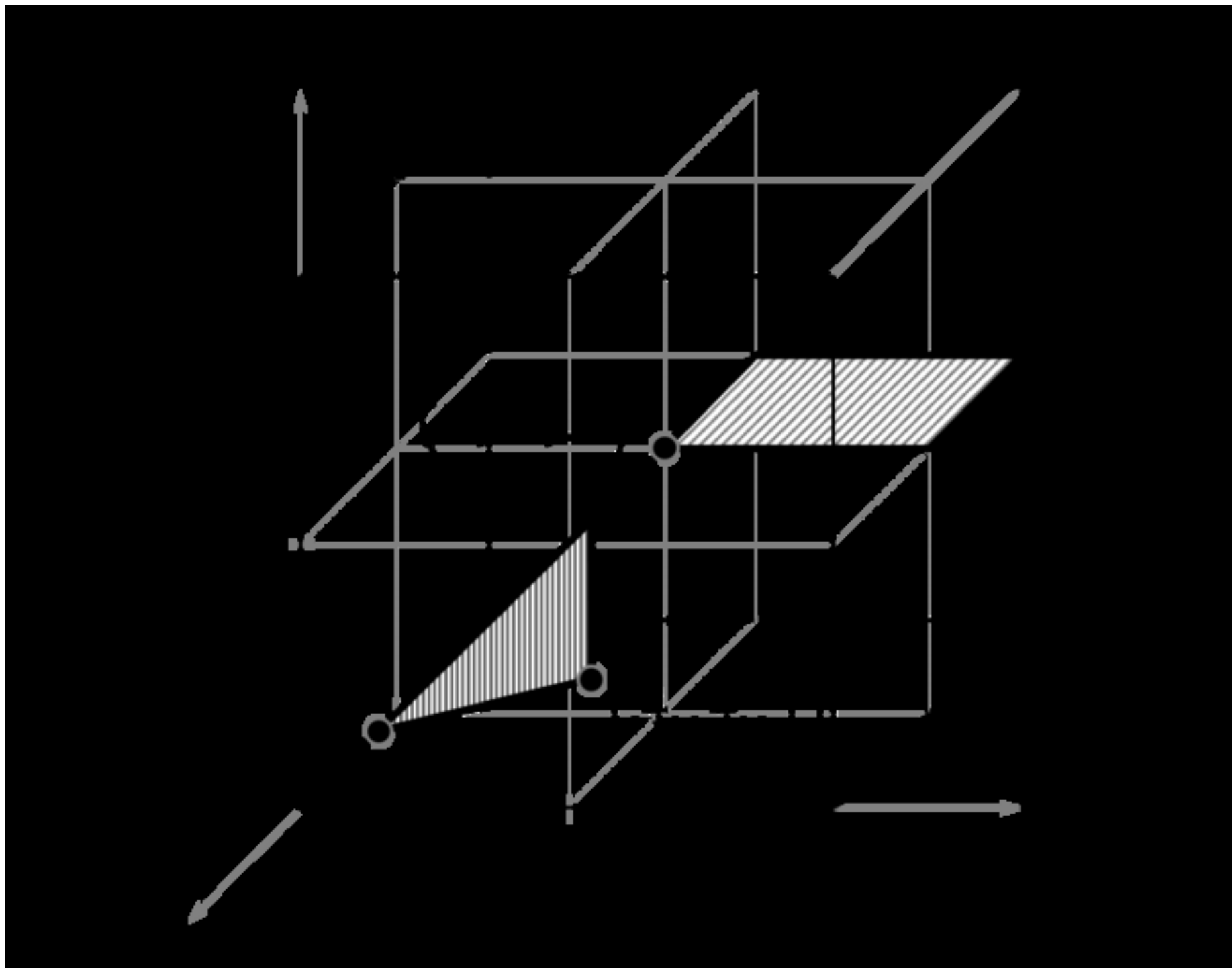
The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:











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